Assignment 4

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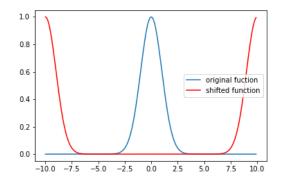
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Q1)

Here is my defined function and choice of gaussian function:

```
def shift(y,dx):
    N=y.size
    kvec=np.arange(N)
    yft=np.fft.fft(y)
    J=np.complex(0,1)
    yft_shift=yft*np.exp((-2*np.pi*J*kvec*dx)/N)
    y_shift=np.fft.ifft(yft_shift)
    return y_shift

x=np.arange(-10,10,0.1)
y=np.exp(-0.5*x**2)
dx=100.0
y_shift=shift(y,dx)
```

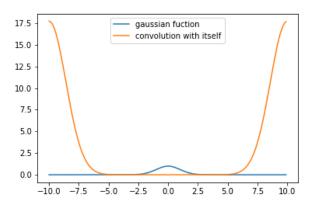


The gaussian is shifted by half of the array length.

Q2)

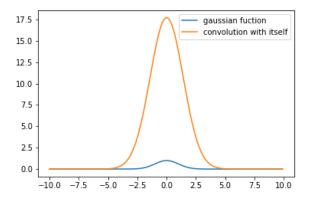
```
def conv(f,g):
    ft_f=np.fft.fft(f)
    ft_g=np.fft.fft(g)
    ft_g=np.conj(ft_g)
    convolution_before_shift=np.fft.ifft(ft_f*ft_g)
    convolution=np.fft.fftshift(convolution_before_shift)
    return convolution, convolution_before_shift
```

The result from **np.fft.fft** is not centered. Considering the periodicity, I used **np.fft.fftshift** to **move** the convolution **to the center**. Here I showed both plots.

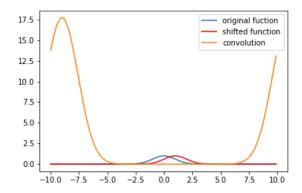


The gaussian I chose is $y = \exp(-0.5*x**2)$. This plot is not easy to interpret.

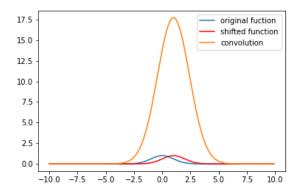
However, when shift the convolution, which is a gaussian, to the center, it makes much more sense now. A **fat gaussian** becomes a **skinny gaussian** after Fourier Transform.



Q3)



Again, I used **np.fft.fftshift** to move the convolution to the center. Comparing with the convolution curve with guassian itself, the convolution curve between gaussian and its shifted gaussian **moves** towards the shifted gaussian.

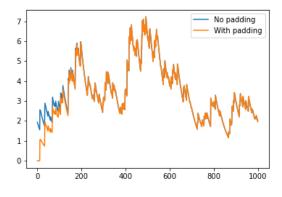


Q4)

```
# NO padding, wrapi-around with danger
f1=0.0*x
for i in range(nhit):
    f1[x_hit[i]]=f1[x_hit[i]]+y_hit[i]
g1=np.exp(-1.0*x/tau)
T1=np.fft.irfft(np.fft.rfft(g1)*np.fft.rfft(f1))
T1=T1[:len(x)]
```

Here I used the example showed in class. Doing convolution with no padding might cause the problem of wrap-around, which is due to the periodicity of Fourier transfer. As you can see from the plot, the curve without padding is trying to connect the starting point and ending point. Padding the data with some zeros and chopping them off in the end will solve this problem.

```
#Padding, wrap-around without danger
f2=0.0*x
f2=np.pad(f2,[0,len(f2)])
for i in range(nhit):
    f2[x_hit[i]]=f2[x_hit[i]]+y_hit[i]
g2=np.exp(-1.0*x/tau)
g2=np.pad(g2,[0,len(g2)])
T2=np.fft.irfft(np.fft.rfft(g2)*np.fft.rfft(f2))
T2=T2[:len(x)]
```



a)

Let $\alpha = exp(-2\pi i k/N)$,

$$\sum_{x=0}^{N-1} exp(-2\pi i k x/N) = \sum_{x=0}^{N-1} \alpha^x = S = 1 + \alpha + \alpha^2 + \dots + \alpha^{N-1}$$

$$S * \alpha = \alpha + \alpha^2 + \dots + \alpha^N$$

$$S - S * \alpha = 1 - \alpha^N$$

$$S * (1 - \alpha) = 1 - \alpha^N$$

$$S = \frac{1 - \alpha^N}{(1 - \alpha)}$$

Bring $\alpha = exp(-2\pi i k/N)$ back to the equation:

$$\sum_{x=0}^{N-1} exp(-2\pi i k x/N) = \frac{1 - exp(-2\pi i k/N)^N}{1 - exp(-2\pi i k/N)}$$
$$= \frac{1 - exp(-2\pi i k/N)}{1 - exp(-2\pi i k/N)}$$

b)

According to L'Hôpital's rule, when k→0:

$$\lim_{k \to 0} \sum_{x=0}^{N-1} exp(-2\pi i k x/N) = \lim_{k \to 0} \frac{1 - exp(-2\pi i k)}{1 - exp(-2\pi i k/N)}$$

$$= \lim_{k \to 0} \frac{(1 - exp(-2\pi i k))'}{1 - exp(-2\pi i k/N)'}$$

$$= \lim_{k \to 0} \frac{2\pi i * exp(-2\pi i k)}{(2\pi i/N) * exp(-2\pi i k/N)}$$

$$= \lim_{k \to 0} \frac{exp(-2\pi i k)}{exp(-2\pi i k)} N$$

$$= \lim_{k \to 0} \frac{exp(0)}{exp(0)} N$$

$$= N$$

When integer k is not a multiple of N, k/N will not be an integer. $exp(-2\pi ik) = 1$,

but
$$\exp\left(-\frac{2\pi ik}{N}\right) \neq 1$$
. $\frac{1 - exp(-2\pi ik) \to 0}{1 - exp\left(-\frac{2\pi ik}{N}\right) \to \neq 0} = 0$.

According to Euler's formula:

$$F(k') = \sum_{x=0}^{N-1} \sin\left(\frac{2\pi i k x}{N}\right) * \exp\left(-2\pi i k' x/N\right)$$

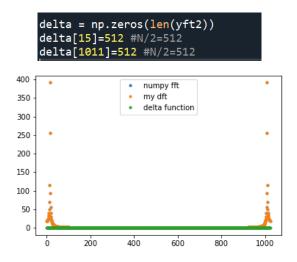
$$= \sum_{x=0}^{N-1} \frac{\exp\left(\frac{-2\pi i k x}{N}\right) - \exp\left(\frac{2\pi i k x}{N}\right)}{2i} * \exp\left(-2\pi i k' x/N\right)$$

$$= \sum_{x=0}^{N-1} \frac{\exp\left(\frac{-2\pi i (k' + k) x}{N}\right)}{2i} - \frac{\exp\left(\frac{-2\pi i (k' - k) x}{N}\right)}{2i}$$

With this analytic estimate of the DFT, I wrote my DFT function as below:

```
def mydft(k_sin,y):
    N = len(y)
    x=np.arange(N)
    kvec = np.fft.fftfreq(N,1/N)
    F = np.zeros(N)
    for i in range(len(kvec)):
        F_i= np.sum(np.exp(-2*np.pi*1j*(kvec[i]-k_sin)*x/N)/2j-np.exp(-2*np.pi*1j*(kvec[i]+k_sin)*x/N)/2j)
        F[i] = abs(F_i)
    return F, kvec
```

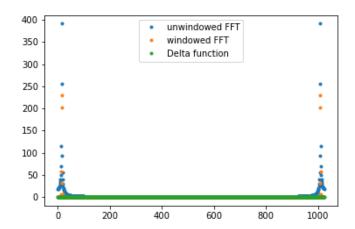
I chose a non-integer k = 15.4, which will reflect as a peak in the Fourier Transform of the sine wave near k'=15. Due to aliasing, there be another peak near k'=1009, which is the mirror image about N/2 of k'=15. This is also consistent with the analytic estimate of DFT. When k'=k or k'=-k, $\exp\left(\frac{-2\pi i(k'+k)x}{N}\right)$ and $\exp\left(\frac{-2\pi i(k'-k)x}{N}\right)$ are not zero and all the other k' will have a DFT equal to zero. Here is how I defined my delta function:



As you can see, the numpy fft(blue) is overlapping with "my dft", but both have some spectral leakage from the delta function. The error between my written DFT and numpy FFT is close to the machine precision, and the error between my written DFT and delta function is huge:

Error between my written DFT and numpy FFT= 9.40319434843738e-13
Error between my written DFT and delta function= 23.63514640992592

d)



With windowing (Orange dots), the spectral leakage effects are reduced. The error between FFT and delta function also dropped.

Error between delta and FFT without window 23.635146409925916 Error between delta and FFT with window 21.599791805286124

e)

$$F(k') = \sum_{x=0}^{N-1} (0.5 - 0.5\cos(2\pi x/N)) * \exp(-2\pi i k' x/N)$$

$$F(k') = \sum_{x=0}^{N-1} (0.5 - 0.5 \frac{\exp(\frac{2\pi i x}{N}) + \exp(-\frac{2\pi i x}{N})}{2}) * \exp(-2\pi i k' x/N)$$

$$F(k') = \sum_{x=0}^{N-1} 0.5 \exp(-\frac{2\pi i k' x}{N}) - 0.5 \frac{\exp(\frac{2\pi i x(k'-1)}{N})}{2} - 0.5 \frac{\exp(-\frac{2\pi i x(k'+1)}{N})}{2}$$

When k'=0, F(0)=N/2,

When k'=1, F(0)=-N/4,

When k' = -1, F(0) = -N/4,

All the other k' will give a zero for DFT. The Fourier Transform of the window is [N/2, -

F(f * window) = F(f)F(window)